

An asynchronous algorithm for massive pseudo-spectral simulations of turbulence on Summit



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Fluid Turbulence: Overview and Challenges

- ► Disorderly fluctuations in time and 3D space, over a wide range of scales (which increases with the Reynolds no.)
- ► A search for scale similarity, with energy cascade from large scales to small scales (via intermediate scales)
- ► Some fluctuations can be extreme (intermittent), requiring better resolution than commonly thought/practiced [1]
- ► Agent of efficient mixing and dispersion, may be coupled to other phenomena (buoyancy, chemical reaction, etc)
- Our emphasis: fundamental understanding in simplified geometries, yet of general relevance to applications

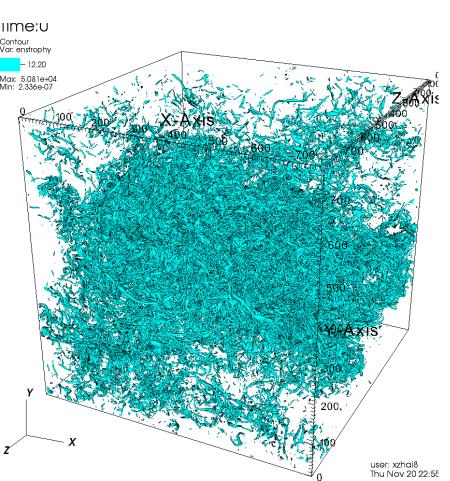


Figure: Vortex filaments in 8192³ direct numerical simulation

Governing equations and numerical methods

▶ Navier Stokes, for conservation of mass and momentum:

$$\nabla \cdot \mathbf{u} = 0 \tag{1}$$

- $\partial \mathbf{u}/\partial t + \mathbf{u} \cdot \nabla \mathbf{u} = -\nabla(p/\rho) + \nu \nabla^2 \mathbf{u} + \mathbf{f}$ (2 Fourier pseudo-spectral methods using FFT for 3D
- domain with periodic boundary conditions
 Grid spacing should be comparable to, preferably smaller than smallest scales (Kolmogorov)
- ▶ 2nd or 4th order Runge-Kutta in time (wavenumber space)
- ► Time step based on Courant number for numerical stability

Major Algorithmic Elements

- ► Domain decomposition: can be 1D (slabs) or 2D (pencils)
- ► One FFT per direction with data local to each MPI process
- ► Transpose using alltoall communication (or variants)
- ► Pack and unpack in local memory before/after alltoall

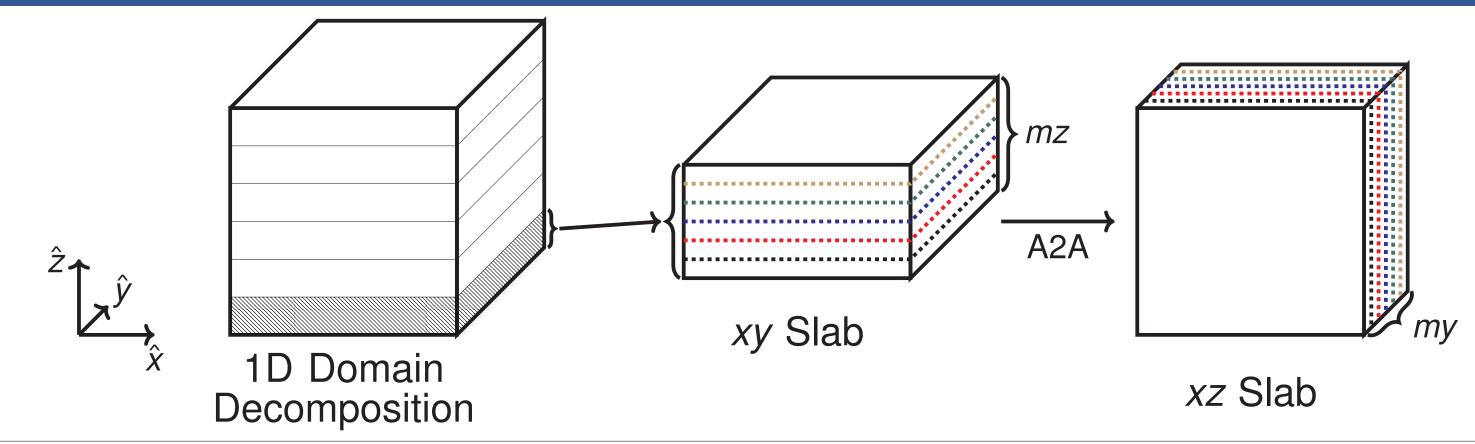
Target on Summit: 16384³ or higher

- ► Fat nodes offer large memory: 1D decomposition
- Fast CPU-GPU data transfer through NVLINK
- ► Spectrum-MPI for communication: 1-sided is best
- Fine-grained overlapping among CPU/GPU computations, NVLINK transfers and non-blocking alltoalls

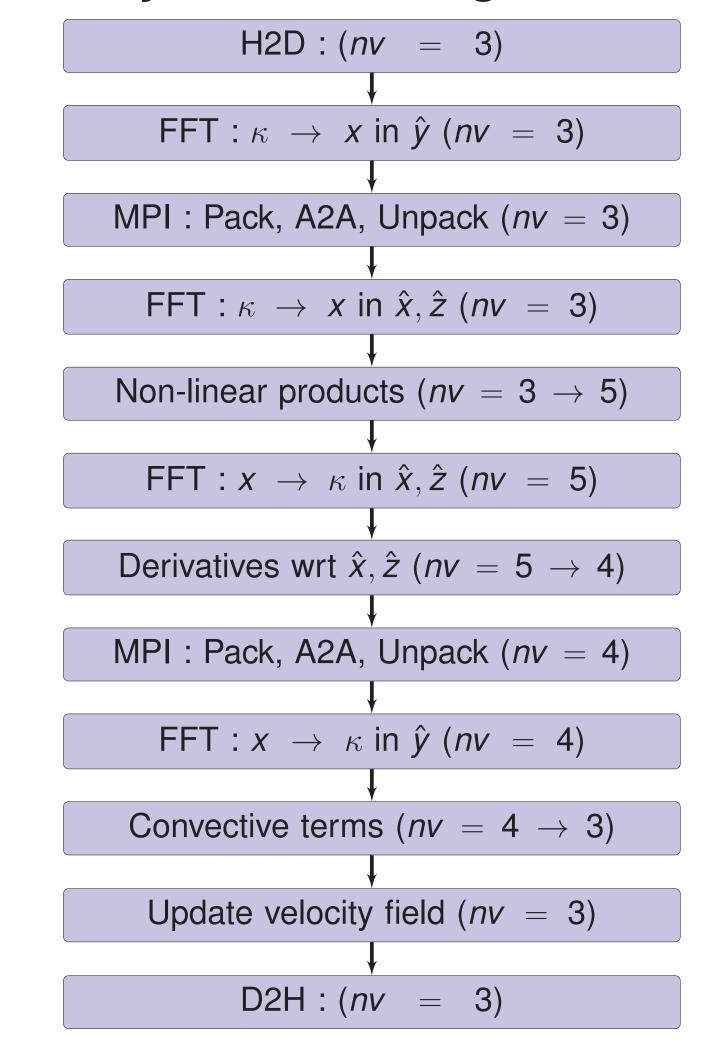
IBM XL compiler, CUDA Fortran[2] for host-device data copies, CUDAFFT on GPUs

- Detailed profiling using NVPROF
- Measurements of network bandwidth
- Small reproducers helped diagnose bugs in system

Domain Decomposition & Algorithm



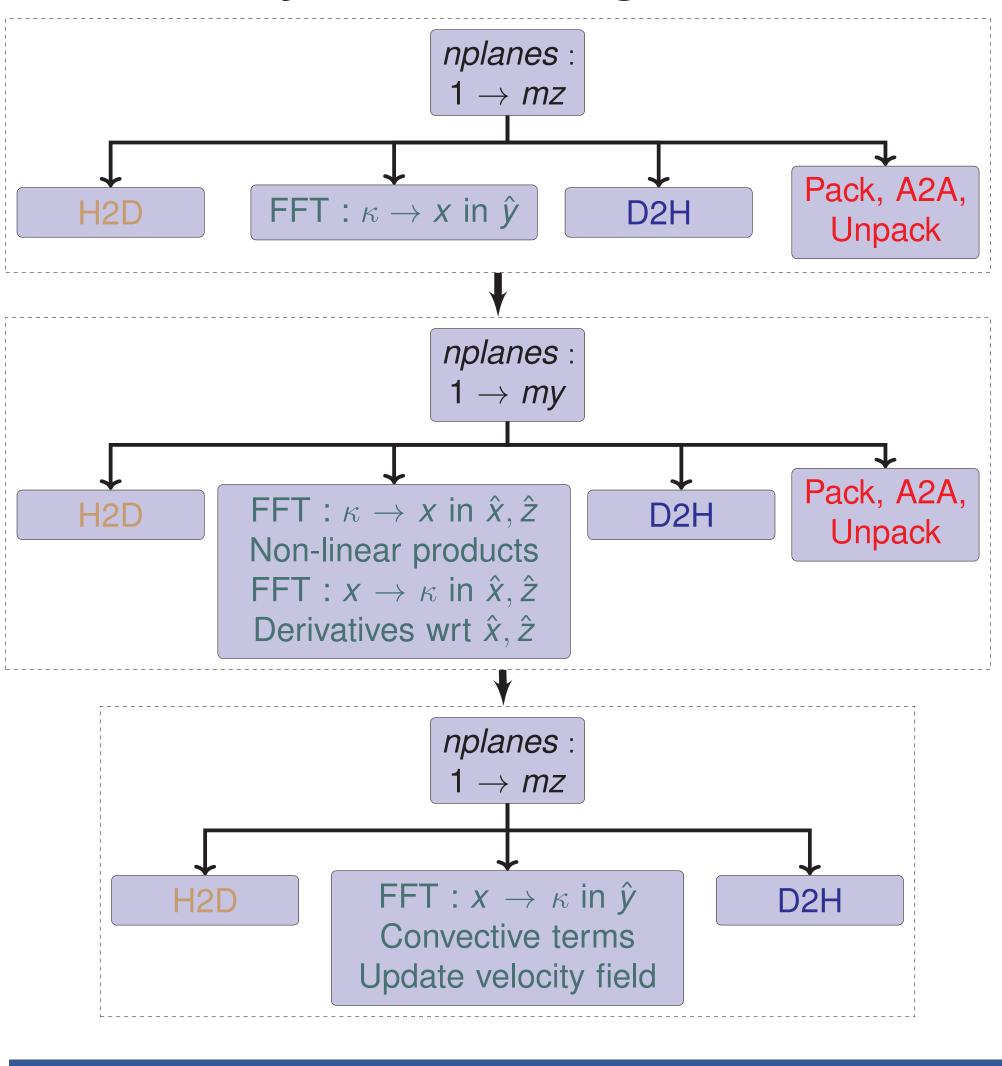
Synchronous Algorithm



Limitations of synchronous algorithm

- Problem size restricted by available GPU memory
- Computations and NVLINK cost add to runtime

Asynchronous Algorithm



Advantages of asynchronous algorithm

- ► Run larger problem using CPU memory
- Overlapping Compute and NVLINK under MPI lowers cost
- ▶ Network, being the bottleneck, is continuously used

Figure: Nvprof timeline of asynchronous code for one Runge-Kutta substep. Alltoall (silver) overlapped with GPU computations and NVLINK (shown in other colors). Further operations have to wait on previous alltoall to complete making it the bottleneck.

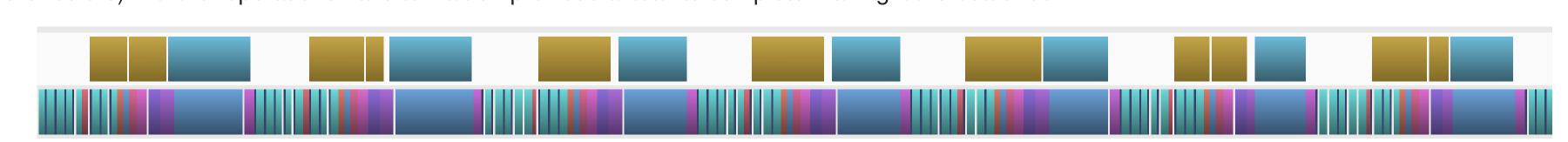


Figure: Overlapping GPU compute and NVLINK data transfers in asynchronous code. First row: Transfer stream (NVLINK), Second row: Compute stream. GPU is continuously used for computations. Data transfers are performed as soon as computations on previous plane are completed

Implementation using CUDA Fortran

- Compute and Transfer streams created
- ► Non-blocking copy between host and device using cudaMemCpyAsync in Transfer stream
- ► cudaFFT and other computes queued in Compute stream
- Synchronization between streams enforced using cudaEventRecord and cudaStreamWaitEvent

Proposed use of OpenMP 4.5

- use_device_ptr clause to call cudaFFT functions
- ► DEPEND and NOWAIT clauses to mimic CUDA streams, events and asynchronous execution

Scaling and Runtime Performance

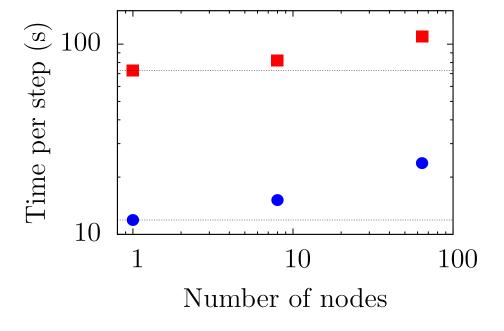


Figure: Scaling of GPU (blue circle) and CPU (red square) code on Summit. Weak scaling of CPU and GPU code for the 64-node problem is 66.35% and 50.23% respectively.

# Nodes	Problem Size	A2A (% total time)	Speedup
1	1536 ³	9.31 (78.16%)	6.11
8	3072 ³	12.40 (81.87%)	5.42
64	6144 ³	20.71 (87.40%)	4.63

Table: Percentage of total time spent on alltoall by the GPU code and speedup of the GPU code compared to the CPU code. Recent modifications to GPU code give speedup > 5X for 1 node problem

Network performance on Summit

- ► Max achieved BW (R+W), 20*GB/s* for larger problems (Theoretical max 46GB/s)
- ► A2A using one-sided MPI tested (gives higher BW)
- ▶ In conversation with IBM spectrumMPI team
- ► Plan to implement hierarchical A2A using non-blocking GATHER and SCATTER

Conclusions & Future Work

- ▶ Process parts of plane instead of full plane at a time
- ► Port cudaFortran code to OpenMP 4.5
- ▶ One-sided MPI for better network performance

Acknowledgments & References

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